



# Landsat-based monitoring protocols for national parks

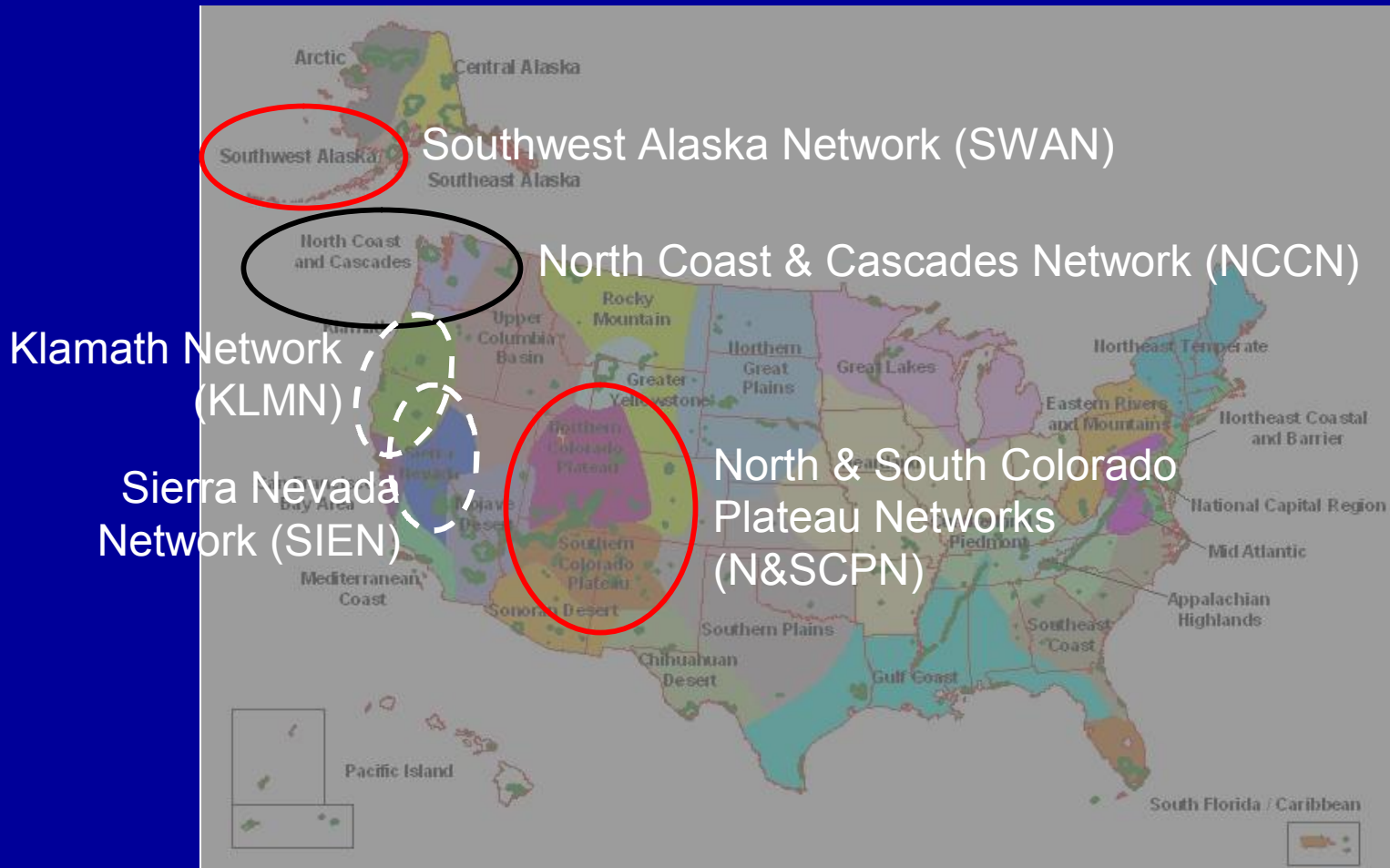
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<http://www.fsl.orst.edu/larse/>

# Monitoring in National Parks

- “Natural Resource Challenge” from Congress forces parks to better understand their systems:
  - Inventory & Monitoring Program
- Key goal in many parks: “Landscape dynamics” and/or “Land cover”
  - *Remote sensing*
- National parks grouped into geographic and ecological units
  - Monitoring networks

# Our NPS collaborations



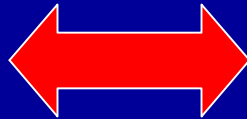
# Remote sensing for monitoring?

- Monitoring → Change detection
- Challenges:
  - Monitor all cover types
    - *Most change detection studies focus on limited types*
  - Monitor all possible change agents
    - *Most change detection studies focus on a few processes*
  - Reference data limited
    - *Ground & airphoto data often collected WITHOUT RS in mind*
  - Multiple user groups for same change product
    - *Outputs must meet needs of scientists, managers, superintendents, and public\*



# A key challenge

“-OLOGIST SPEAK”:  
Ecologists, Geologists,  
Botanists, etc.



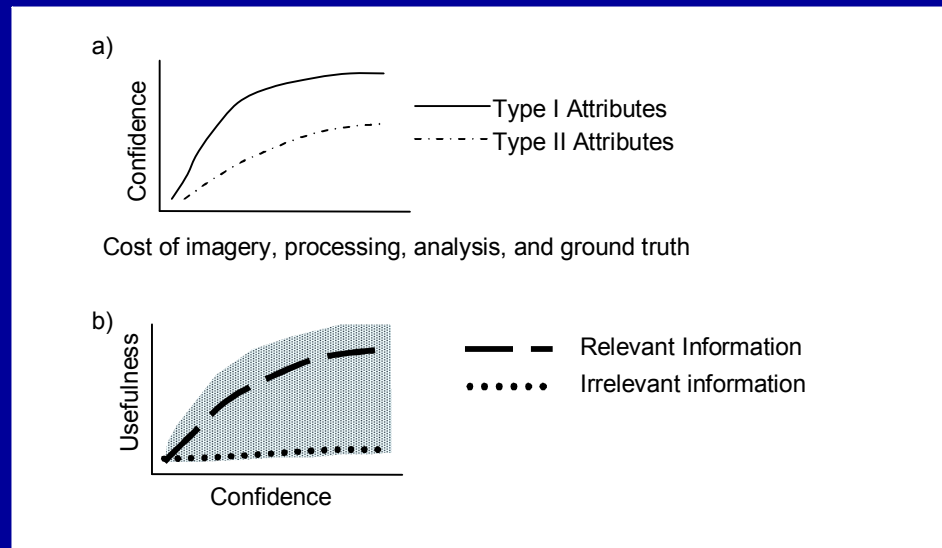
“RS-SPEAK”:  
Remote sensing geeks

# Reframing goals and solutions

- Cost/Confidence/Utility
- Iterative process
- A question of communication

Detecting a  
given  
attribute

But is that  
attribute  
useful?



- Questions re-framed → Novel solutions

# Focal questions

- What process do you want to monitor?
- What are the manifestations of that process on the landscape?
- At what spatial grain do we need measurements to detect those manifestations?
- How often do we need measurements to capture the process?

# Focal questions

- Are the manifestations of desired processes spectrally separable from background?
- Are there sensors that can capture that spectral separation?
- Does the separation rise above levels of background noise over time?



# Monitoring goals framed for remote sensing

Table 1.

Ecological monitoring goals of the NCCN Parks evaluated January 14th and May 13th, 2004, in Seattle, WA. All goals are characterized in terms of spatial and temporal grain. Based on spatial and temporal grain, as well as importance to the NCCN Parks, each goal was assigned a priority for consideration in the study plan. Those that are also likely to be achievable using Landsat-based satellite data are noted.

<i>Topic</i>	<i>Sub-topic</i>	<i>Spatial Grain</i>	<i>Temporal Grain</i>	<i>Priority</i>	<i>Achievable monitoring goal?</i>
Alpine Vegetation	Bare ground impacts	1m	5 y	Skip (need higher resolution)	
	Interface w/forest	1m / 30m	Decadal	High/Advise	YES
	Vegetation Comm.	1m	> Annual	High	Advise
Forest Vegetation	Hardwood/Conifer	30m	> Annual	High	YES
	Forest Structure (classes)	1m / 30m	> Annual	High	YES
Disturbance	Vegetation disturbance in avalanche chutes	1m / 30m	5-10 yrs	High	YES
	Landslides	1m / 30m	Annual / > Annual	High	YES
	Fire	30m	Annual / > Annual	High	YES
	Insect/Disease	1m / 30m	Annual / > Annual	High	YES
	Windthrow	1m/ 30m	Annual / > Annual	High	YES
	Pollution	?	?	Low (important in future; impacts are not extensive enough to detect at present)	

ETC....

# Monitoring goals grouped temporally

- When and where are these occurring?
  - Use remotely-sensed product as alarm and as measurement tools
- Frequency of monitoring affects
  - Methods of analysis
  - Methods of validation

Table 2. NCCN

Monitoring goals  
grouped by change  
interval needed for  
detection

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*Type 1: Monitor yearly*

Avalanche chute clearing

Landslides

Fire

Insect/disease defoliation  
in forest

Windthrow

Riparian disturbance

Clearcuts

Rural development

*Type 2: Monitor decadal*

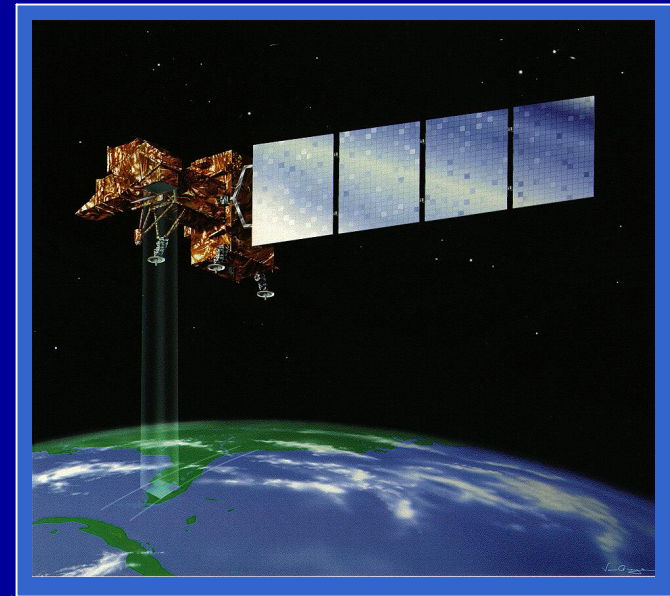
Alpine tree encroachment

Hardwood/conifer forest  
composition

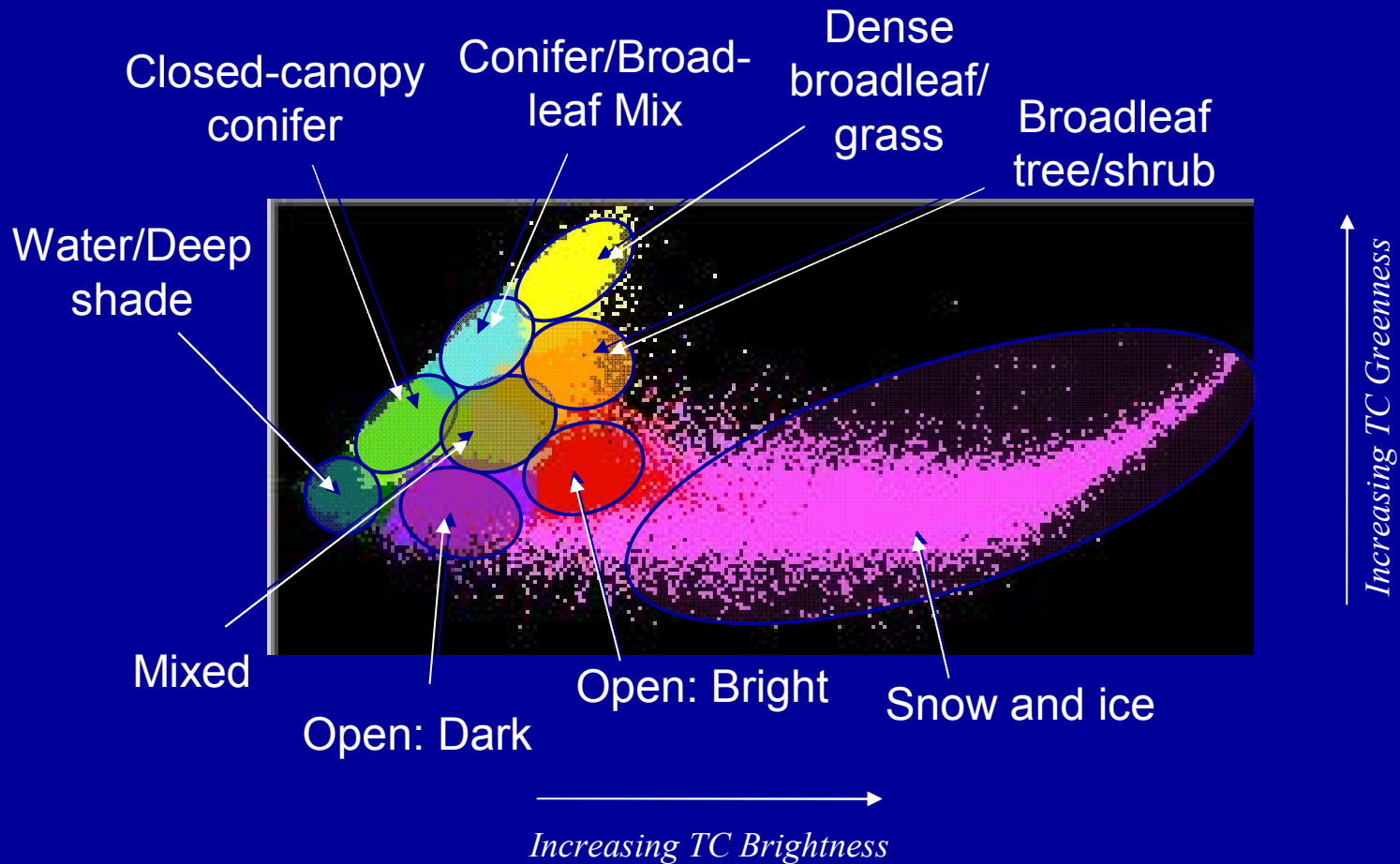
Forest structure

# NCCN: Landsat

- Decision to use Landsat Thematic mapper (TM)
  - Has potential to meet many monitoring goals cheaply and effectively
  - “Eighty percent at half the cost”?
- Characteristics:
  - 1984-present
  - Spatial grain: ~30m
  - Six spectral bands

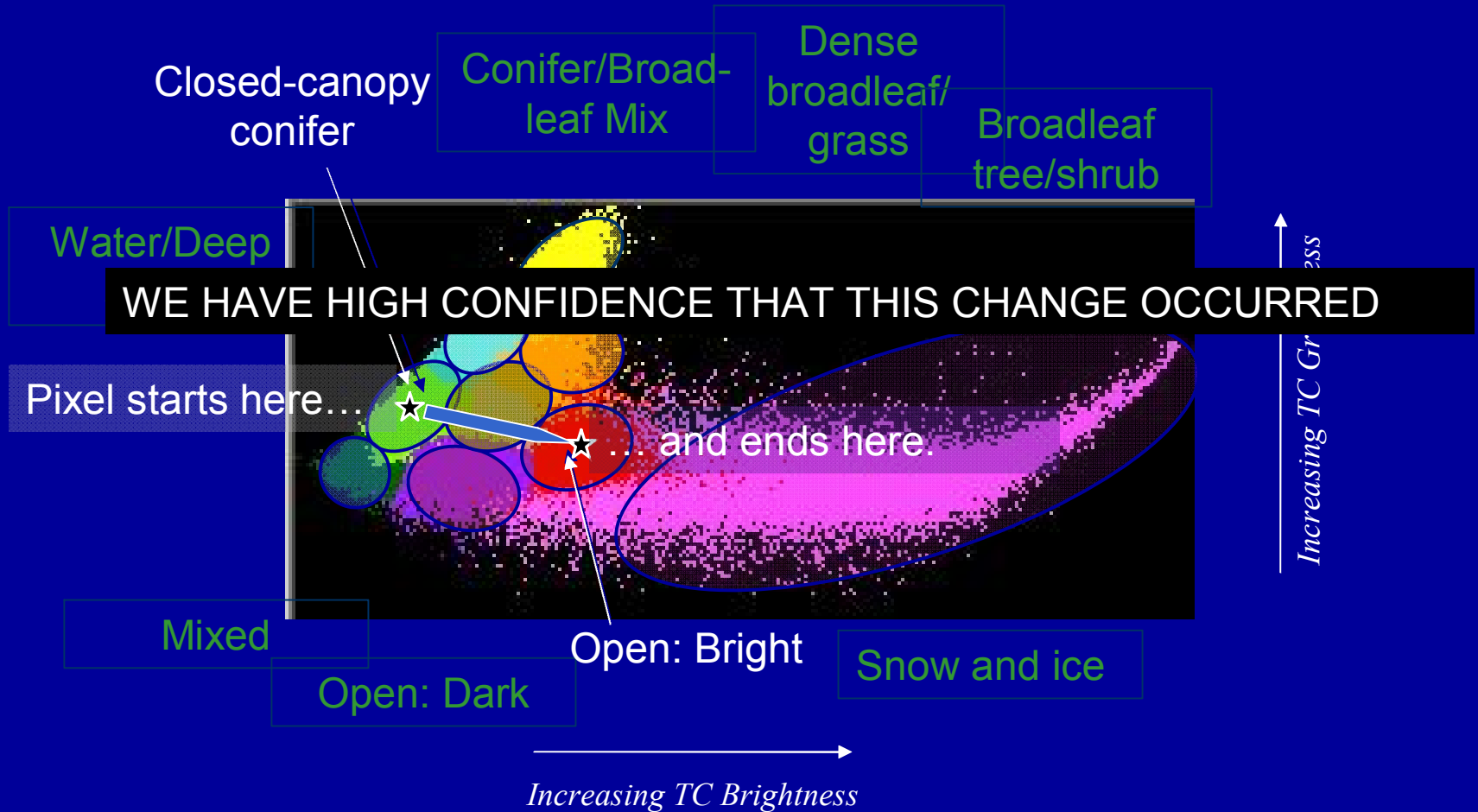


# Physiognomic classes



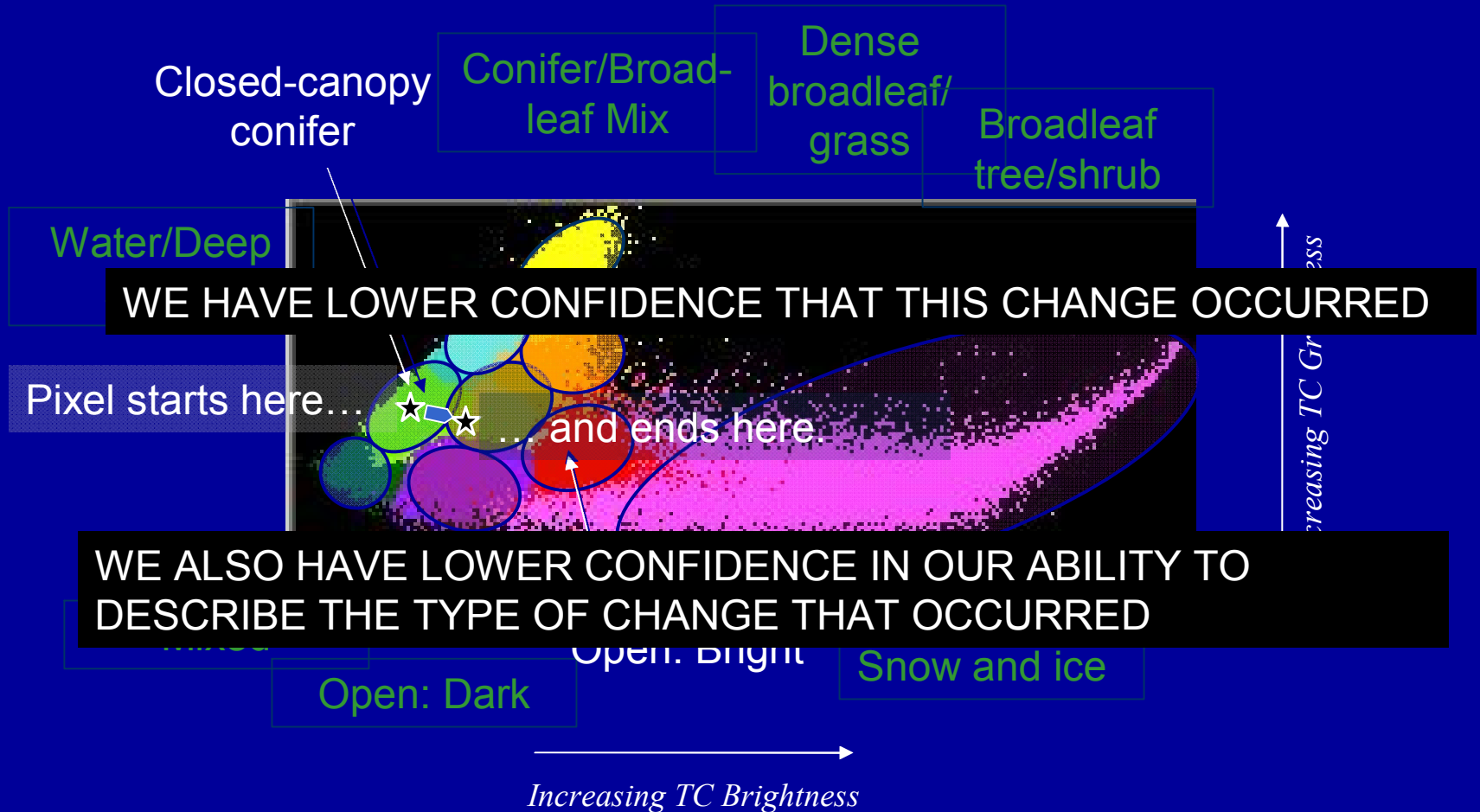
From SOP #3

# Change in POM



From SOP #3

# Change in POM

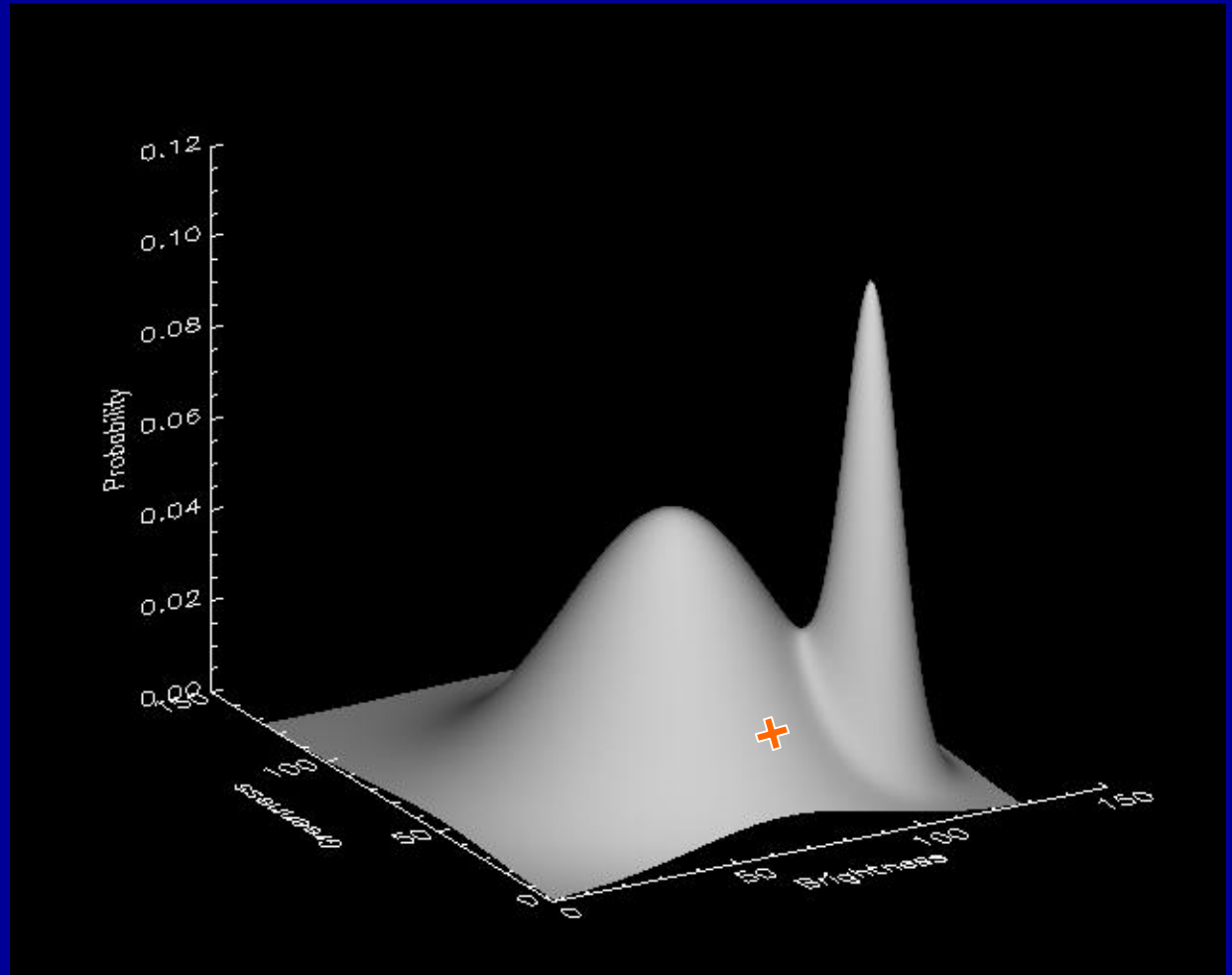


From SOP #3



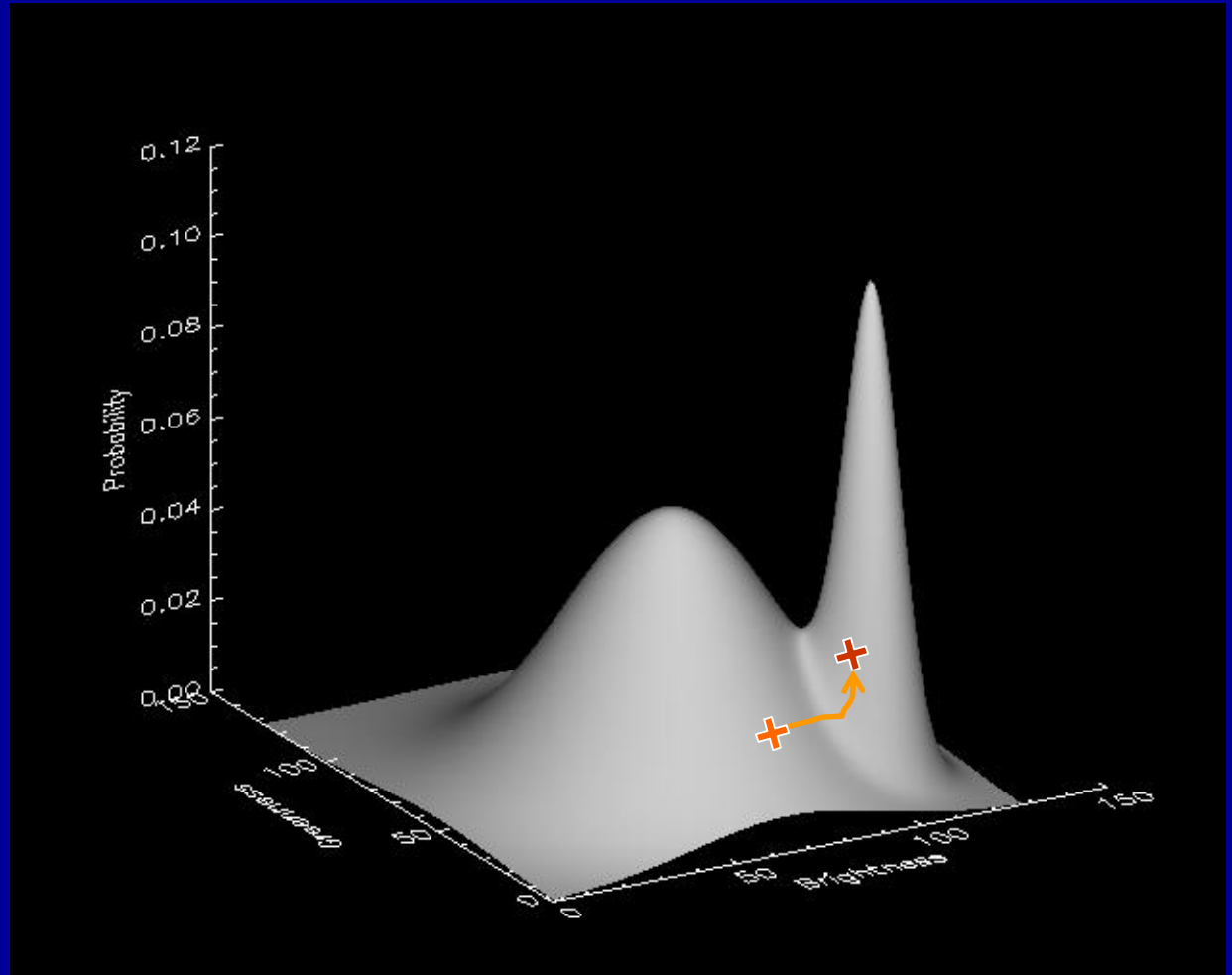
# Change in POM

- Consider Classes 1 and 2
- Date 1: likelihood noted with “+”



# Change in POM

- Combination of “FROM” and “TO” POM changes helps label change
- Magnitude of combined change tells us about distance in probability space
- Apply this to all physiognomic classes and identify greatest changes



# Labeling change

- Subtract probabilities of membership across all classes over time
  - Increases and decreases in probabilities of membership
- Identify largest “from-to” change pair to describe change
- Note absolute value of change – in units of “likelihoods”, akin to probabilities

# Difference in POM

- Adjustment of probability threshold allows separation of easy from difficult change goals

## *Insect mortality near Mt. Rainier*

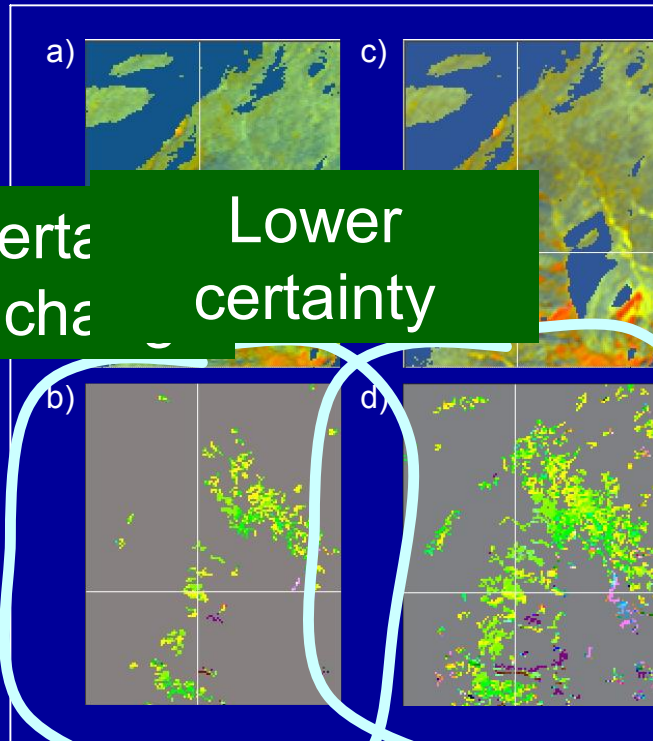


Figure 1. Tasseled cap images and associated change products for an area adjacent to Mount Rainier National Park. a) Tasseled-cap imagery from 1996. Aspect classes are processed separately; this imagery shows only northwest aspects. Tasseled-cap brightness is shown in red, greenness in green, and wetness in blue. Conifer forest appears as light cyan to dark blue, broadleaf vegetation as yellow, and open areas as red or orange. b) The product of the fuzzy change detection approach for the 1996 to 2002 period, showing only areas where probability of membership (POM) in a given class has changed by more than 70%. Red, green, and blue color guns correspond to bare soil, broadleaf, and conifer physiognomic types. Insect mortality results in negative conifer values and/or broadleaf values, leaving yellow (G+R color guns positive) or green (only G color gun positive) tones. c) As in part a, but for the year 2002. d) As in part b, but for a change threshold of 50% rather than 70%.

# Ecological monitoring

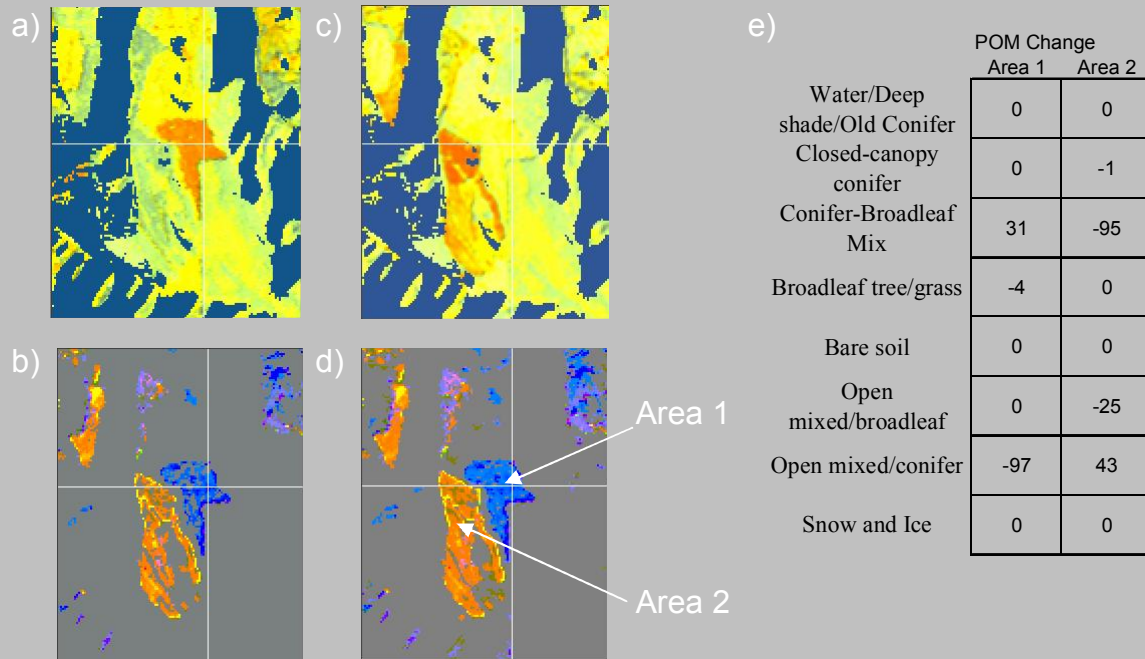
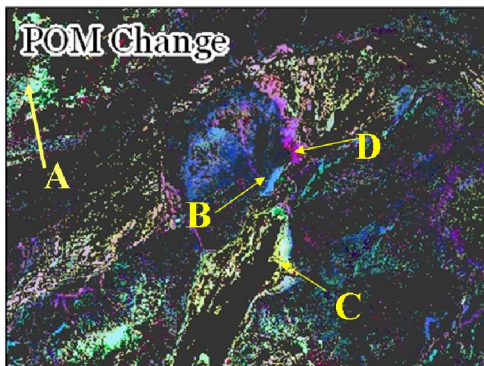
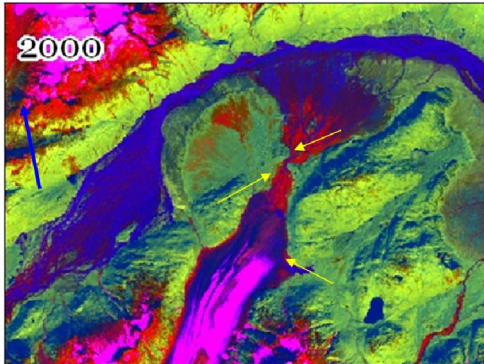
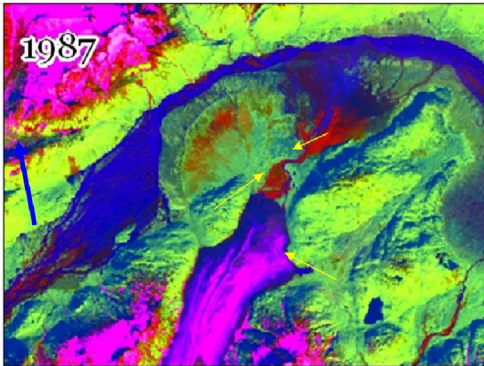


Figure 2. (a-d) As in Figure 1, but for a different area near the park, and with RGB color guns in b) and d) corresponding to open mixed/conifer, open mixed/broadleaf, and conifer-broadleaf mix, respectively. e) Scores for pixels in the two areas shown in part d). Changes are differences in probability of membership for each of the eight physiognomic classes listed on the left-hand column of the table. The directional movement in POM allows interpretation of the changes occurring on the surface directly from the change product alone.



### Tuxedni glacier environs



### Change labels:

A: Snow → Gravel/Sparsely vegetated

B: Gravel/Sparsely vegetated →  
Willow shrub

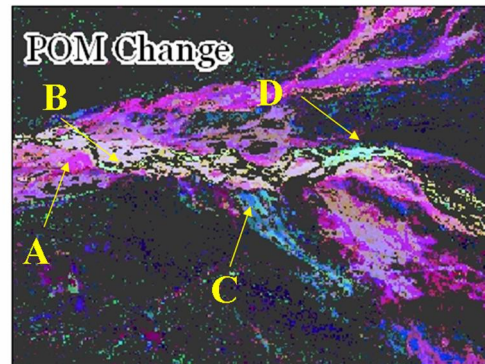
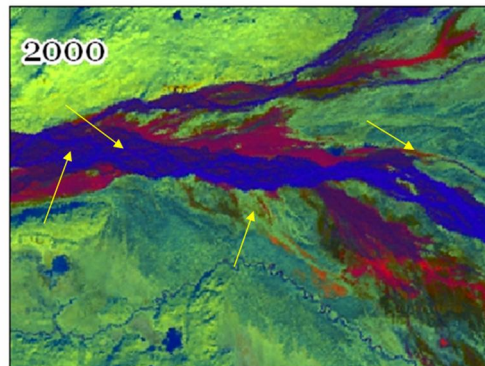
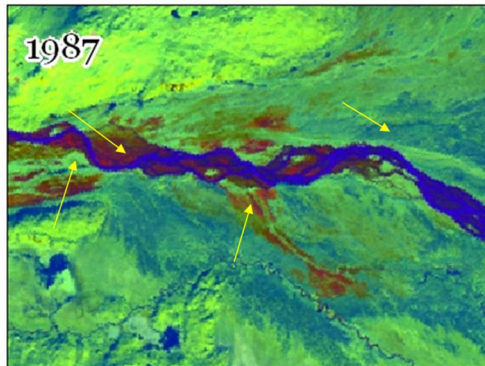
C: Snow → Silty water/Dirty snow

D: Willow shrub → Gravel/Sparsely  
vegetated





## Drift River



Change labels:

A: Snow → Gravel/Sparsely vegetated

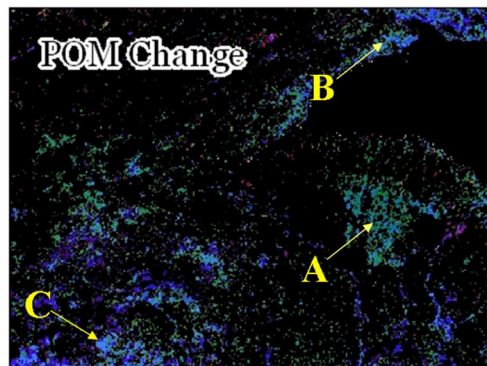
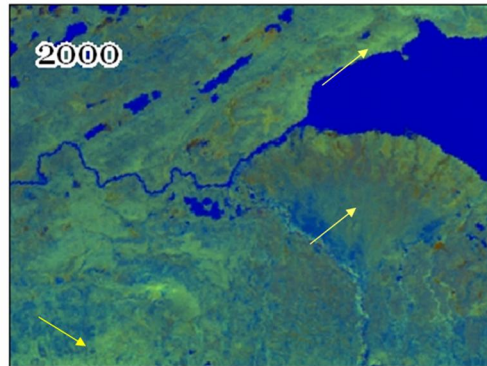
B: Gravel/Sparsely vegetated → Willow shrub

C: Snow → Silty water/Dirty snow

D: Willow shrub → Gravel/Sparsely vegetated



## Telaquana Lake



Change labels:

A: Prostrate shrub tundra → Spruce woodland

B: Sedge meadow → Closed alder

C: Sedge meadow → Closed alder



# Summary

- Frame monitoring goals in terms of remote sensing
- Process is important
- Validation considered from the beginning
- Consider solutions that are
  - Workable
  - 80% of the goal at half the cost



A scenic photograph of a mountain range with snow-capped peaks and a lake in the foreground. The text "The end" is overlaid in the center.

# The end

# Extras

# Reference data

- Ideal reference data:
  - Independent measurement tool
  - At appropriate spatial grain
  - Distributed across entire landscape
  - Acquired at the same time as imagery
  - Repeated as often as repeat imagery for change detection
- Such data do not exist



# Overview of protocol

- Specify detail on ordering and pre-processing of imagery
- Identify physiognomic classes for a baseline image year
- Derive gaussian probability surfaces for physiognomic classes
  - Probability of membership (POM)
- Physiognomic classes overlap in spectral space

# Focus on change

- Develop change maps without reference to baseline map
  - Contrasts with N&SCPN and SWAN
    - Baseline mapping w/airphotos
    - Vegetation mapping program of NPS
      - <http://biology.usgs.gov/npsveg/>
- Frame questions in terms of broad “physiognomic classes”
  - Based on general knowledge of spectral data space

# Validation

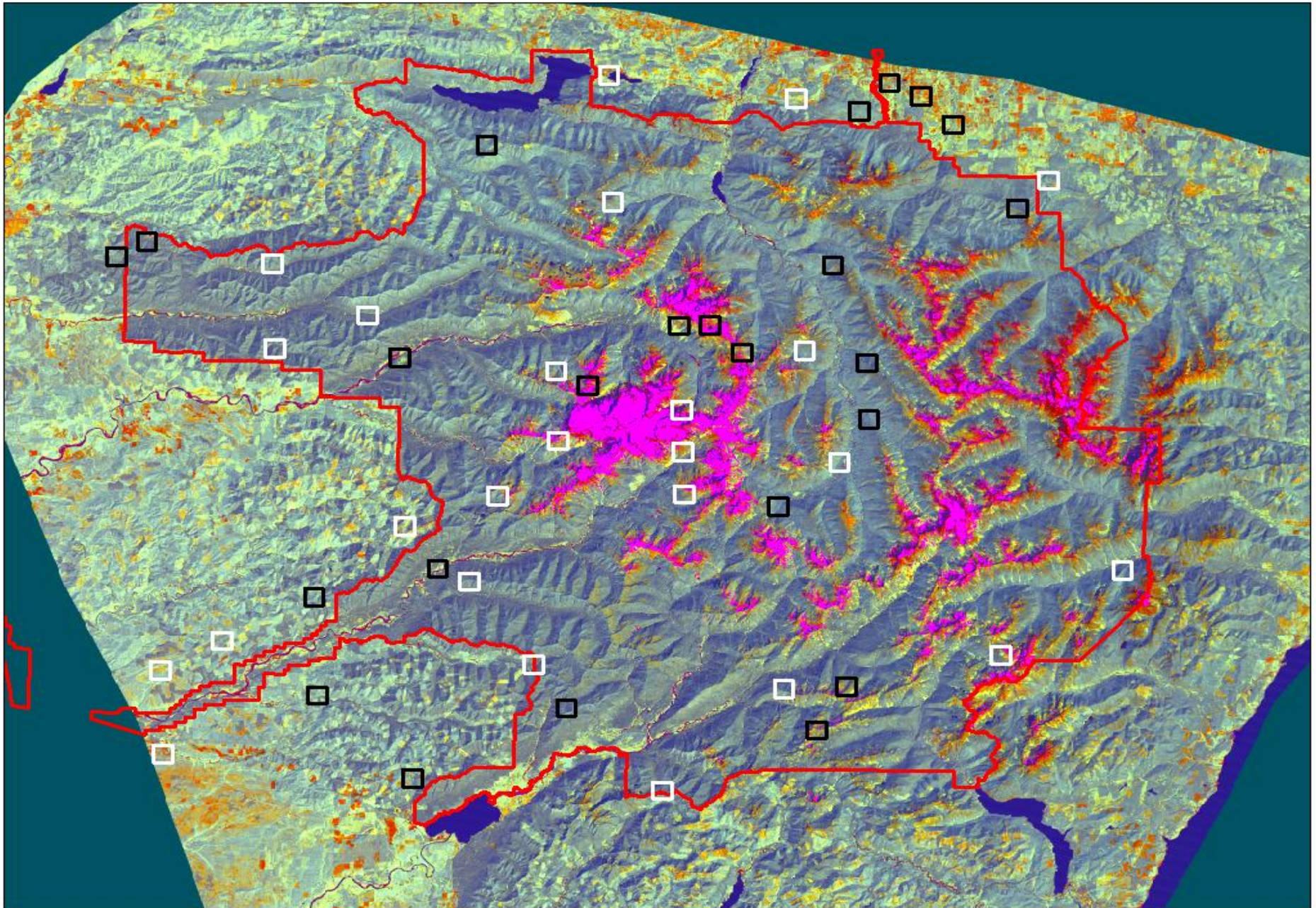
- Type 1 monitoring (disturbances)
  - Primary approach: direct interpretation of satellite imagery (S2S)
    - On-screen digitization
    - Use rules similar to airphoto interpretation – i.e. shape, size, color, texture of events
    - Assign change type, change agent, and confidence in change
  - Test: Essentially how well the difference in POM approach models the human interpreter

# Validation

- Type 2 monitoring (encroachment, revegetation)
  - Airphoto-based interpretation
  - Decadal time-step
  - Use this validation to understand accuracy of S2S validation



Testing in white, training in black





# Regrowth at OLYM

Polygons digitized  
directly onto  
imagery

2002 TC

DOQ

DOQs and TC

- In addition to labeling change type, label disturbance, and ascribe “certainty score” to both calls

Attributes of olym\_trn\_pht\_25pt

	disturbance_agent	certscore_change_type	certscore_dist_agent	
<input type="checkbox"/>	7	4	2	diff image says increase in conif
<input type="checkbox"/>	7	4	2	diff image says increase in conif
<input type="checkbox"/>	7	4	2	diff image says increase in conif, some black pixels
<input type="checkbox"/>	7	4	2	diff image says increase in conif, some black pixels
<input type="checkbox"/>	2	3	2	not sure, high elevation, could be smoothing
<input type="checkbox"/>	8	3	0	unknown, possible phenology

Record: 65 Show: All Selected Records (0 out of 93 Selected.) Options



## S2S Disturbance Agent Types

S2S change number	S2S change type
1	no change
2	water to rock/soil
3	water to partial veg cover
4	water to complete veg cover
5	rock/soil to water
6	partial veg to water
7	complete veg to water
8	increase in broadleaf
9	increase in conifer
10	increase in broadleaf and conifer
11	decrease in broadleaf
12	decrease in conifer
13	decrease in conifer and broadleaf
14	increase in snow
15	decrease in snow

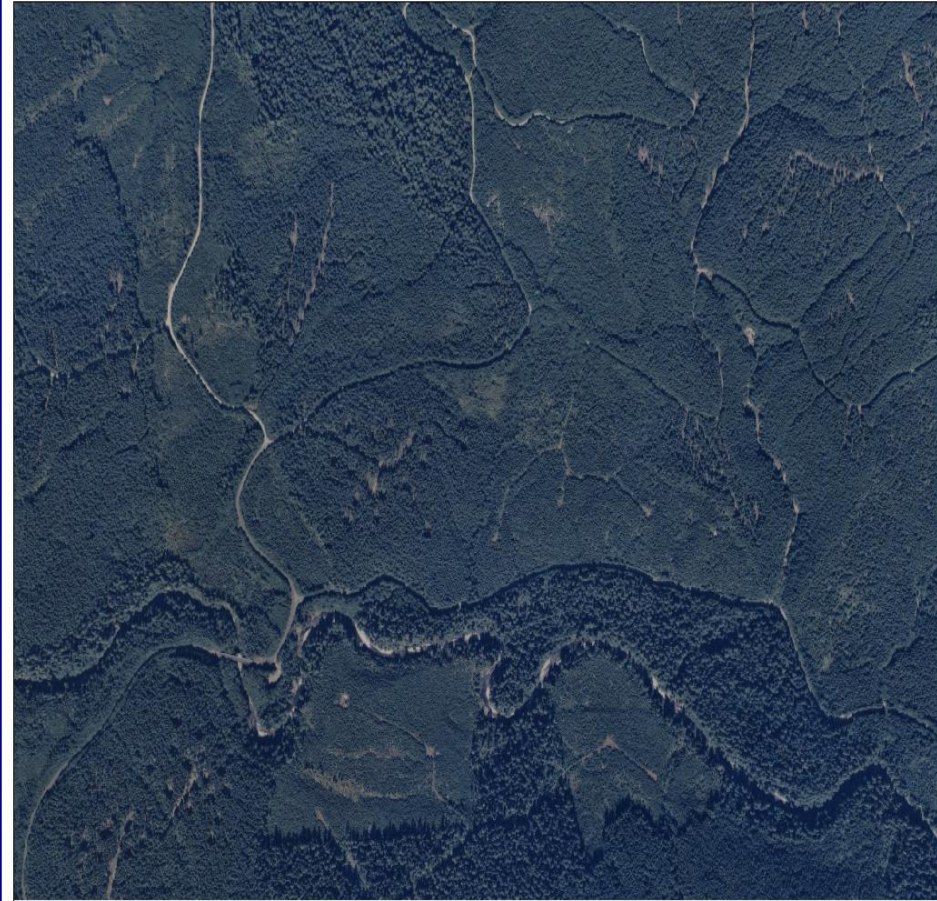
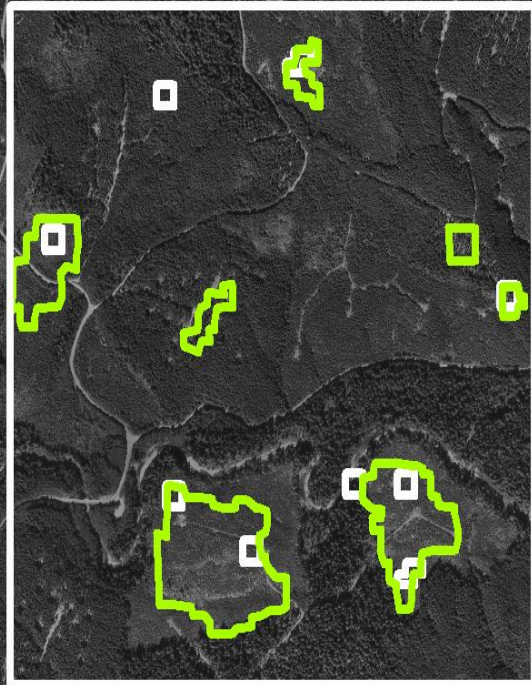
# Satellite validation

- Testing S2S (satellite-satellite) validation
  - Sample S2S polygons with .25 ha plots
  - Use 2000 DOQs and 2002 scanned airphotos to make change calls
    - Double-blind

# Regrowth at OLYM

DOQ

Scanned airphoto

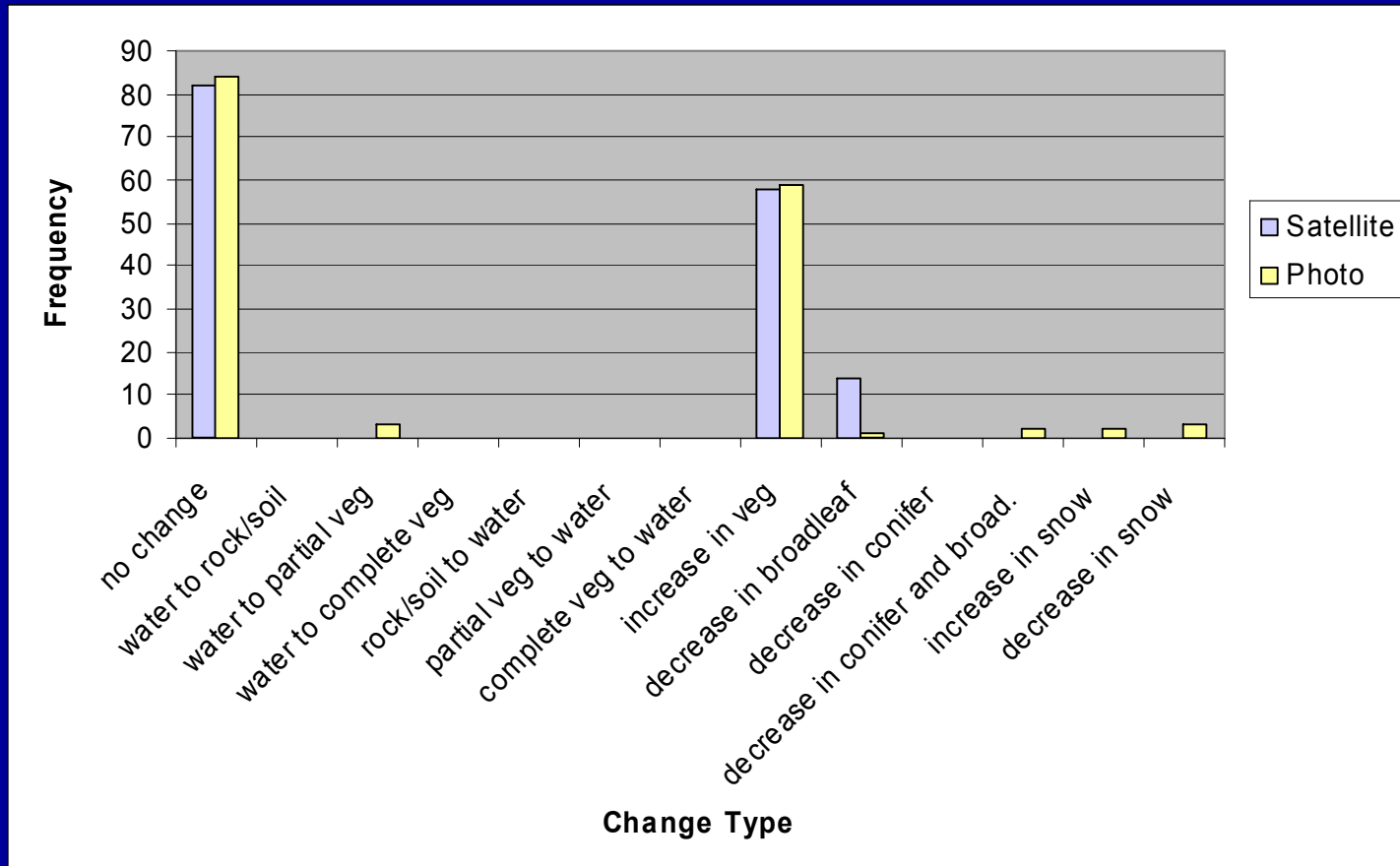


Attributes of olym\_trn\_pht\_25pt

	disturbance_agent	certscore_change_type	certscore_dist_agent	
	7	4	2	diff image says increase in conif
	7	4	2	diff image says increase in conif
	7	4	2	diff image says increase in conif, some black pixels
	7	4	2	diff image says increase in conif, some black pixels
	2	3	2	not sure, high elevation, could be smoothing
	8	3	0	unknown, possible phenolou

Record: 65 Show: All Selected Records (0 out of 93 Selected.) Options

# S2S and Air Photo Comparison with vegetation increase change types collapsed into one category



# Validation

- Third level of validation: field
  - Cost prohibitive
  - Protocol defines options for opportunistic, on-the-ground “yea/nay” validation
    - Piggy-back on to other field efforts
  - Sampling design a key concern



# Accuracy Assessment: Error matrix

- Summarize using an error matrix

	Class types determined from reference source				
Class types determine d from classified map	# Plots	Conife r	Hardwoo d	Water	Totals
	Conifer	50	10	2	62
	Hardwoo d	7	20	0	27
	Water	3	0	8	11
	Totals	60	30	10	100

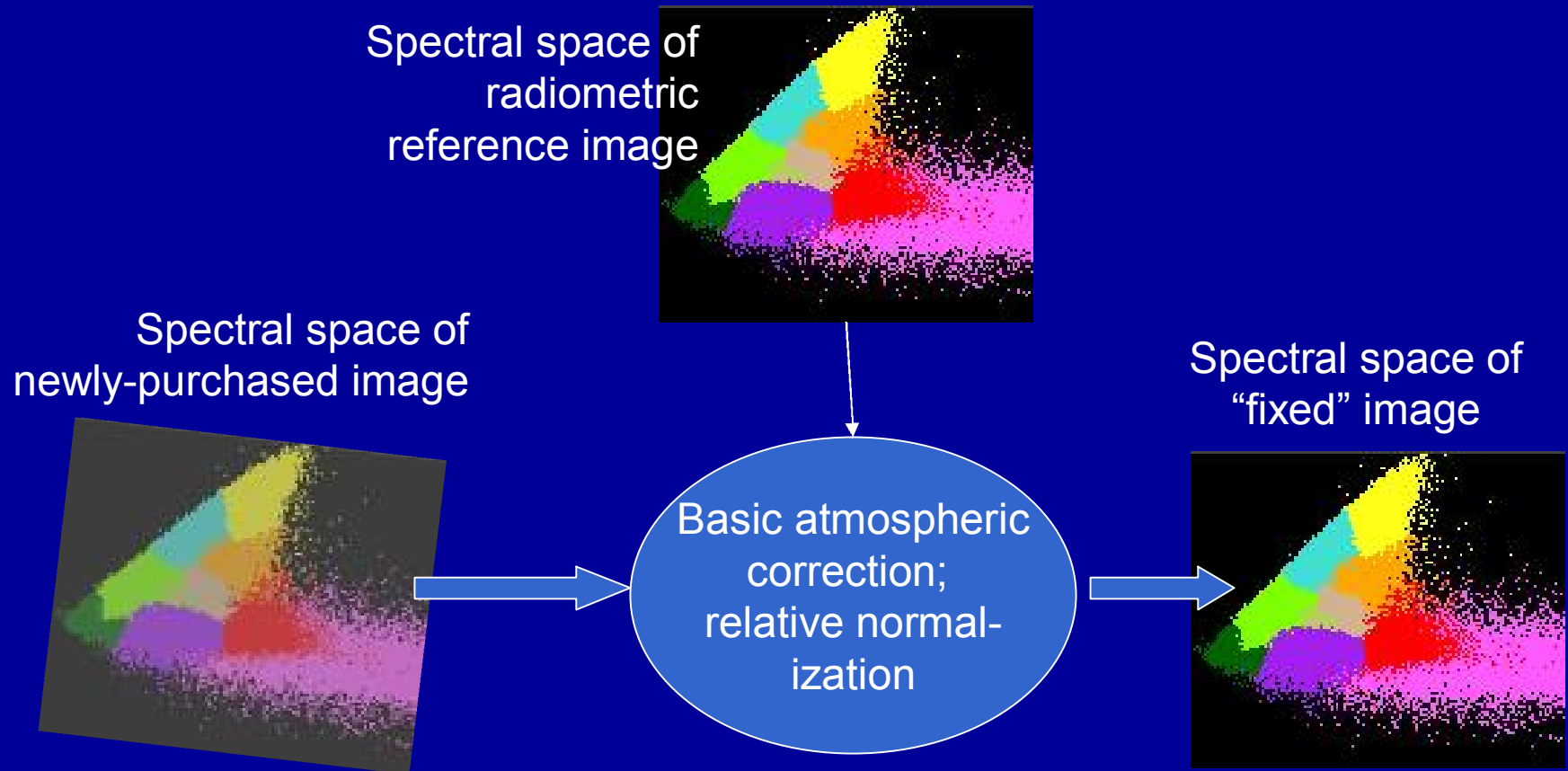


# Accuracy Assessment: *Kappa*

- *Kappa* statistic
- Estimated as  $\hat{K}$
- Reflects the difference between actual agreement and the agreement expected by chance

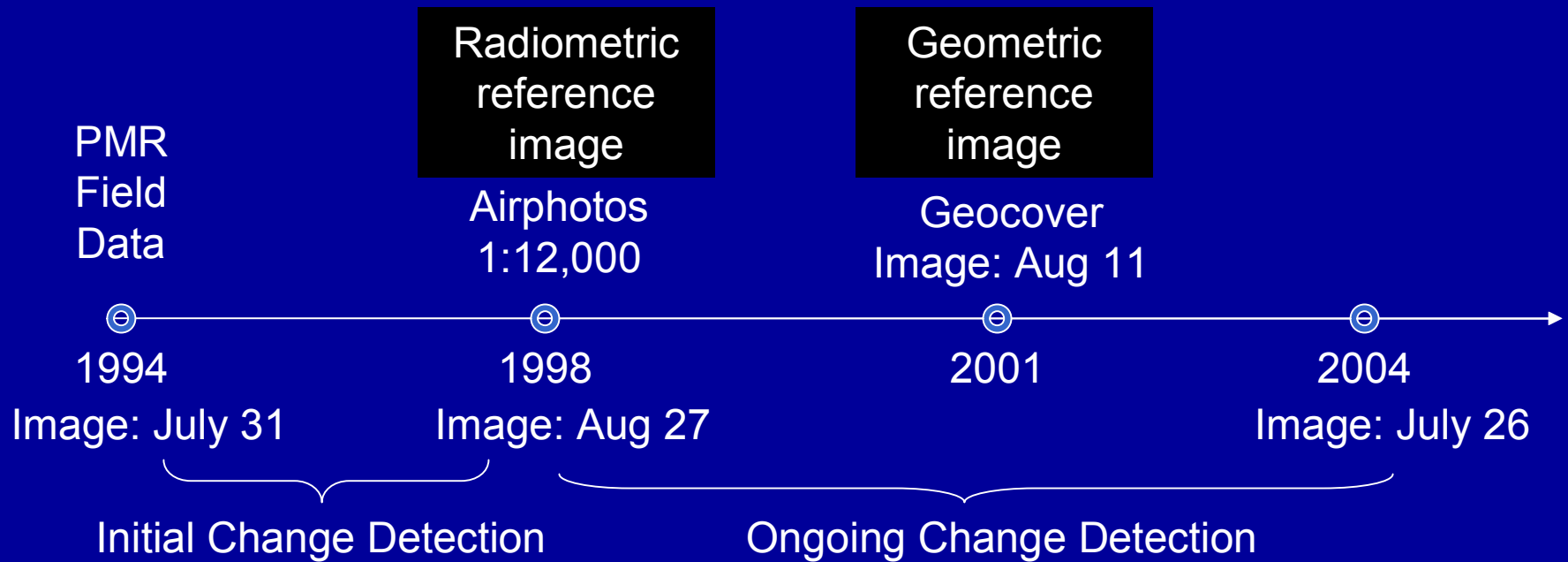
$$\hat{K} = \frac{\text{observed accuracy} - \text{chance agreement}}{1 - \text{chance agreement}}$$

# Radiometric correction



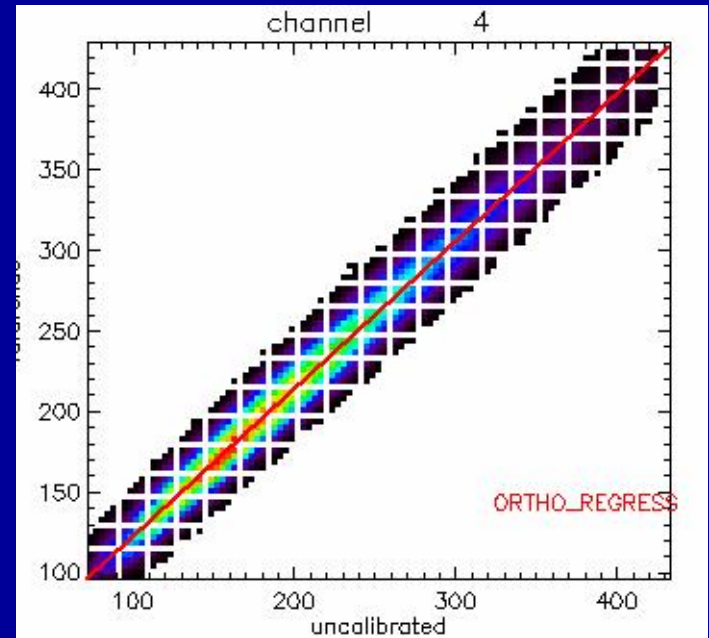
# Overview of methodology

EXAMPLE: NOCA



# Relative normalization

- MADCAL
  - Use “multivariate alteration detection” to identify pixels that are stable across all bands for a reference image and a subject image
  - Develop regressions on a *band by band* basis to map subject image numbers into reference image



## Certainty scoring

### **Change type: 0-5**

Spectral change vector is distinct from change vector of similar starting types in surrounding area 0, 1, or 2

Area of spectral change is large and consistent within “patch”: 0 or 1

Change is clearly not caused by misregistration: 0 or 1

Spectral condition of endpoints is interpretable and consistent with change call: 0 or 1

### **Disturbance agent: 0-3**

Shape is consistent with disturbance agent: 0 or 1

Size is consistent with disturbance agent: 0 or 1

Landscape position and context is consistent with disturbance agent: 0 or 1



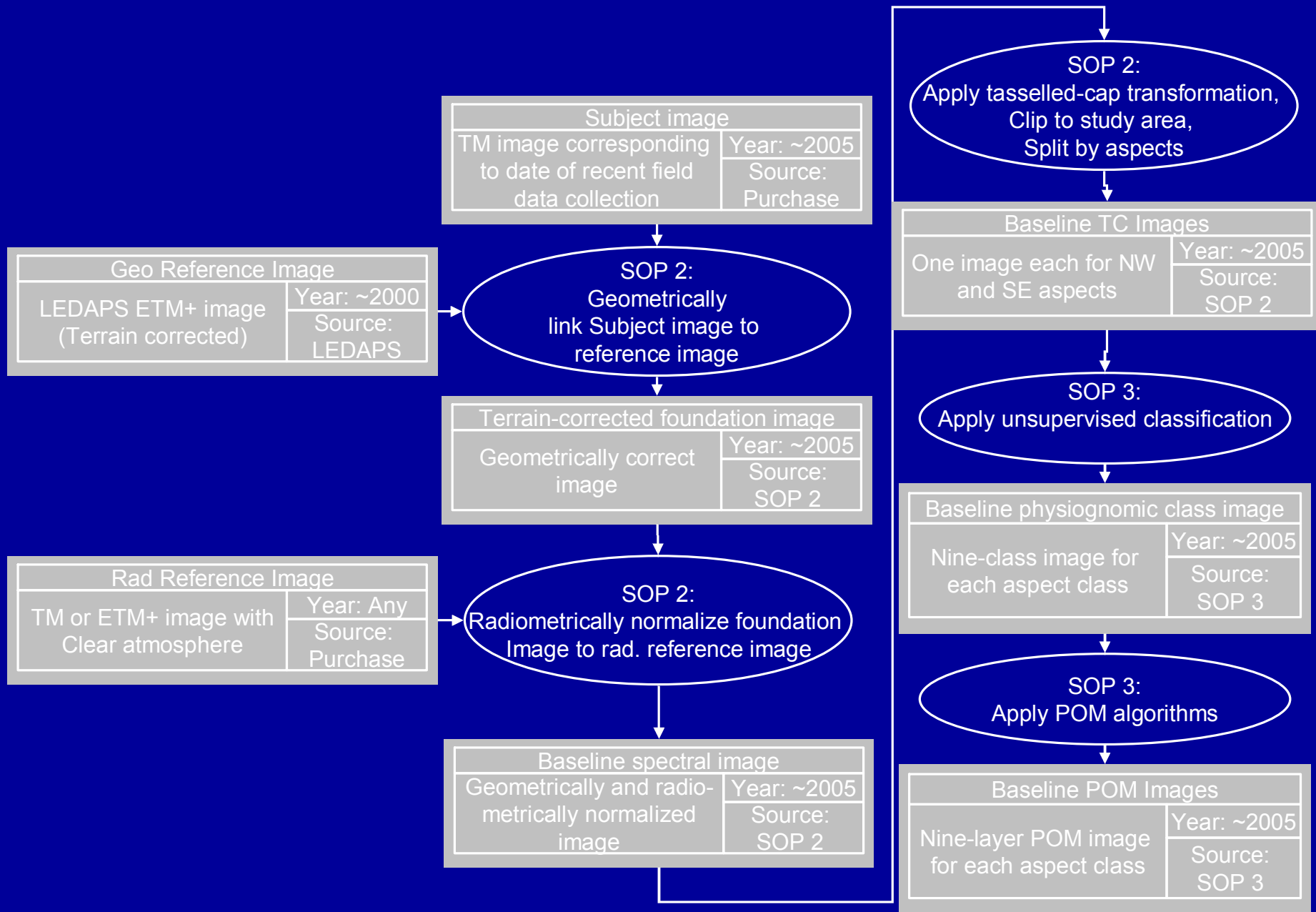


Figure 5. Startup phase.

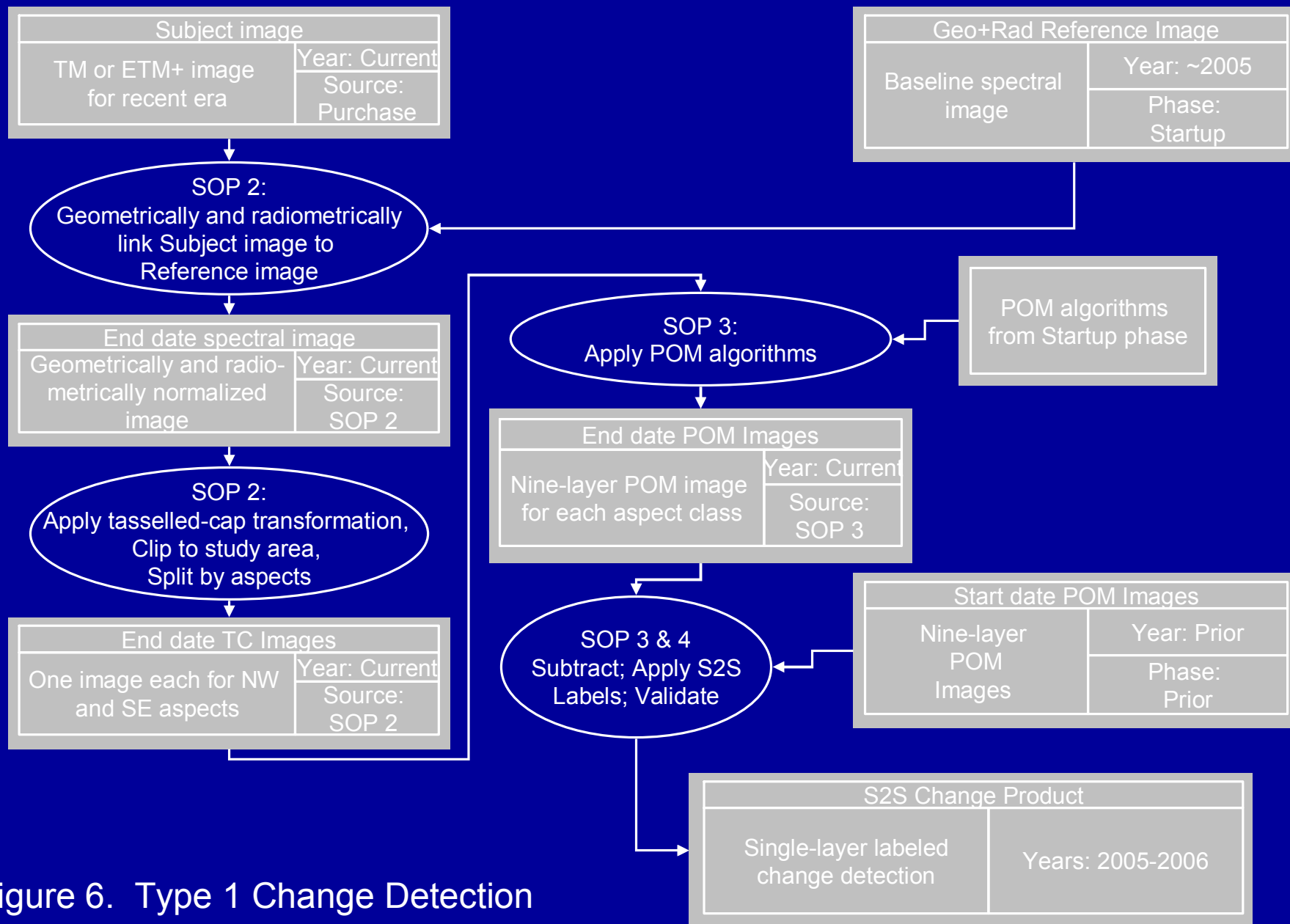


Figure 6. Type 1 Change Detection

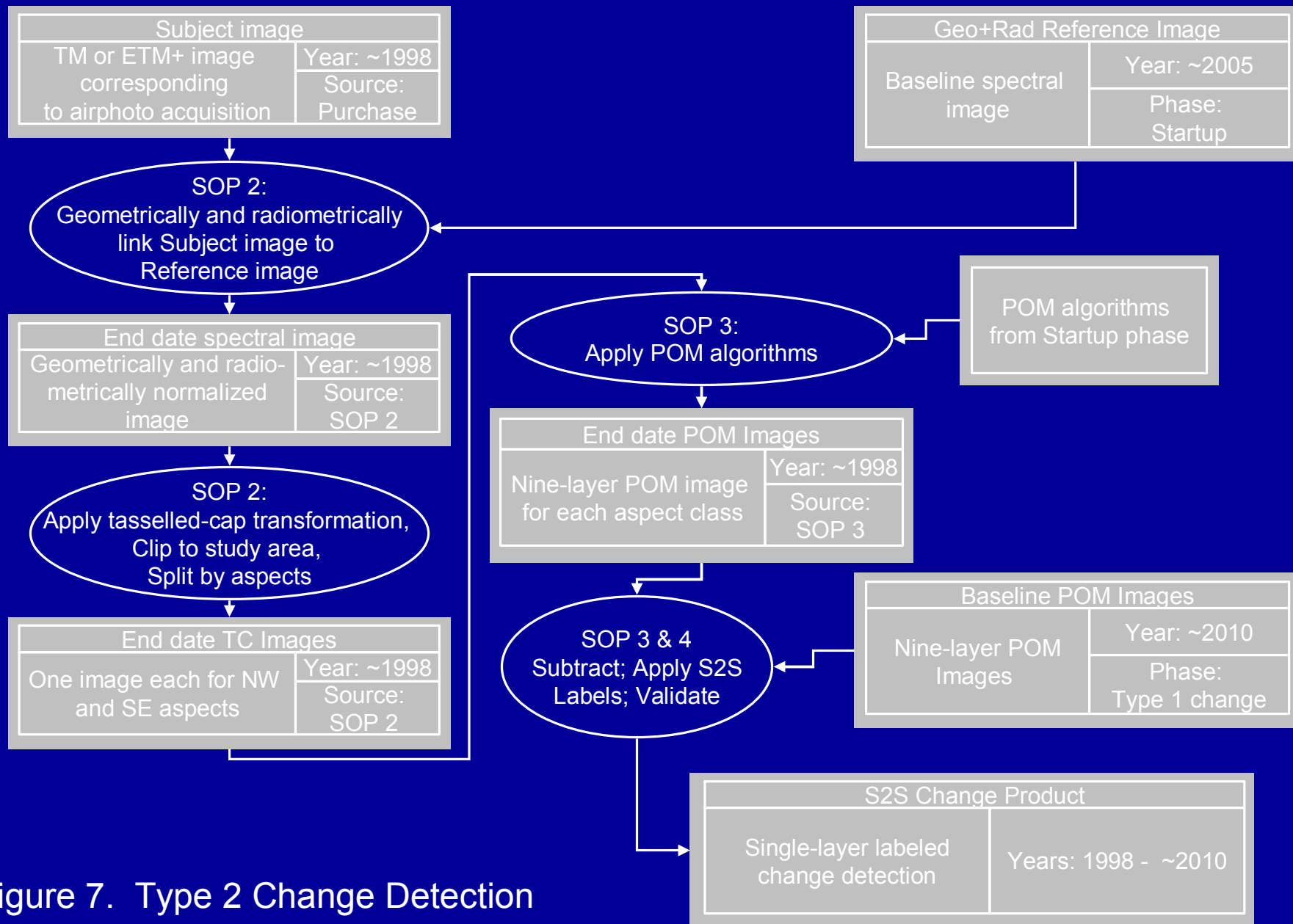
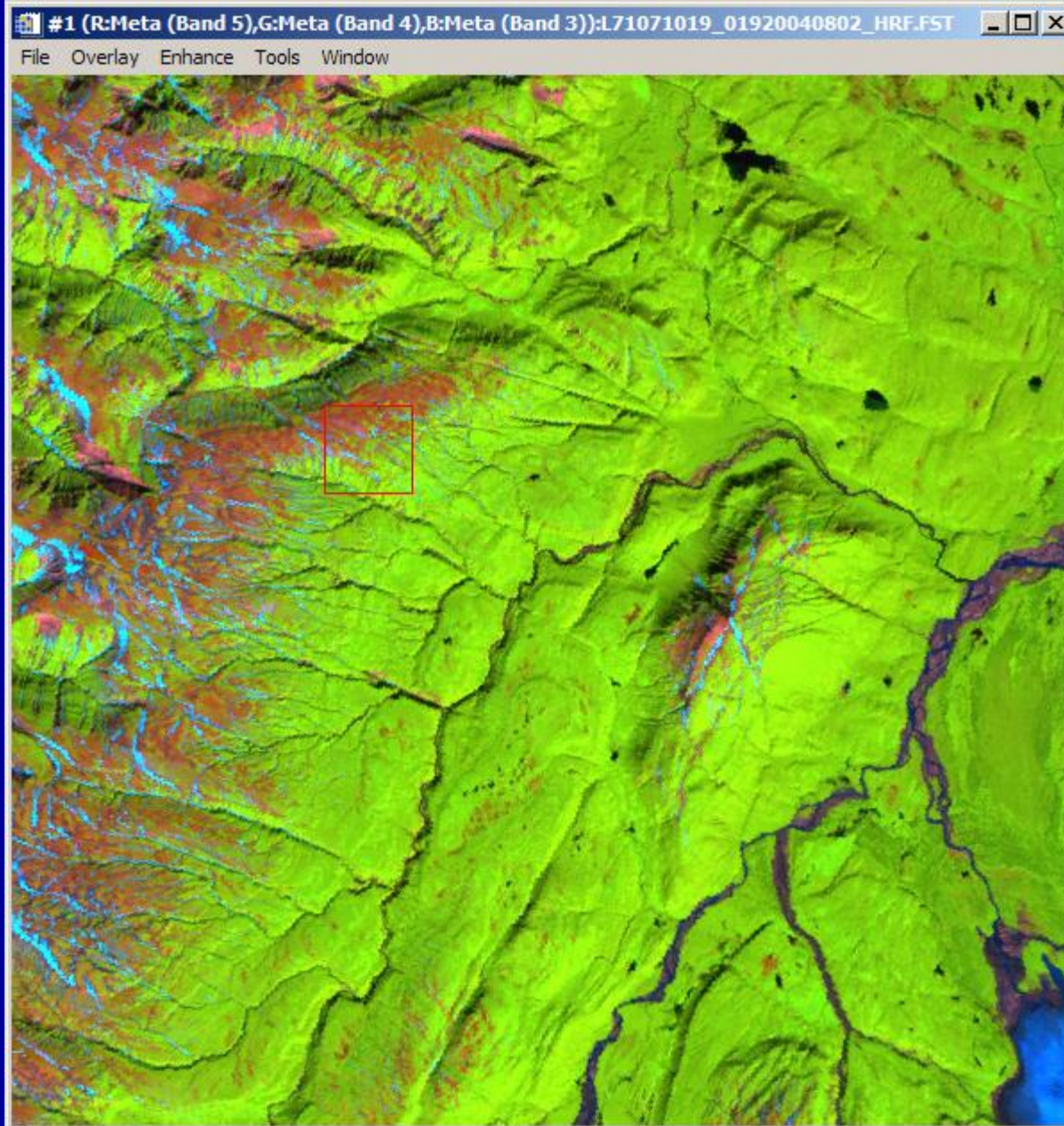
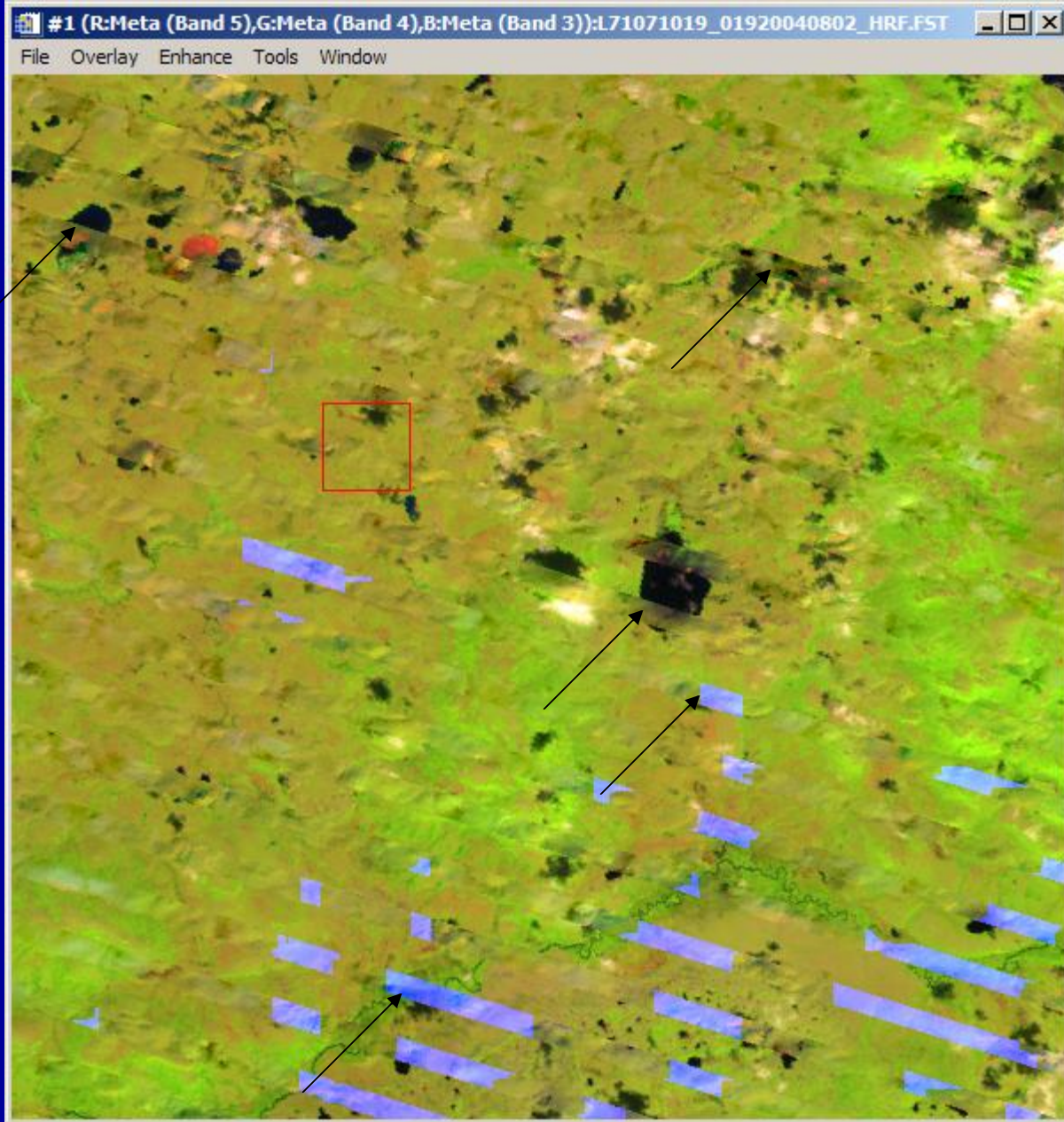


Figure 7. Type 2 Change Detection



Ex 1: No  
noticeable  
problems  
with gap-  
filling

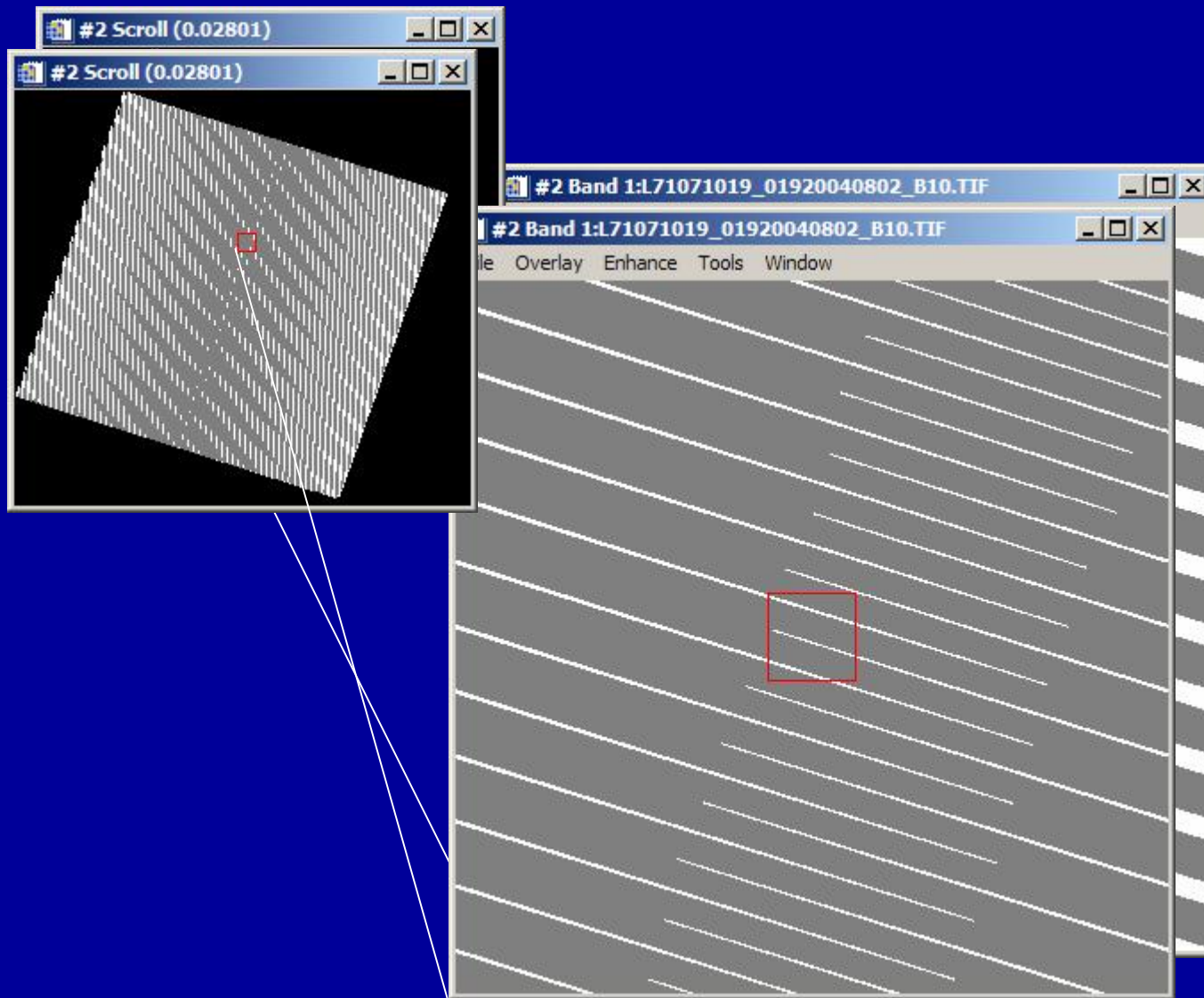




Ex 2: Quite noticeable problems with gap-filling!

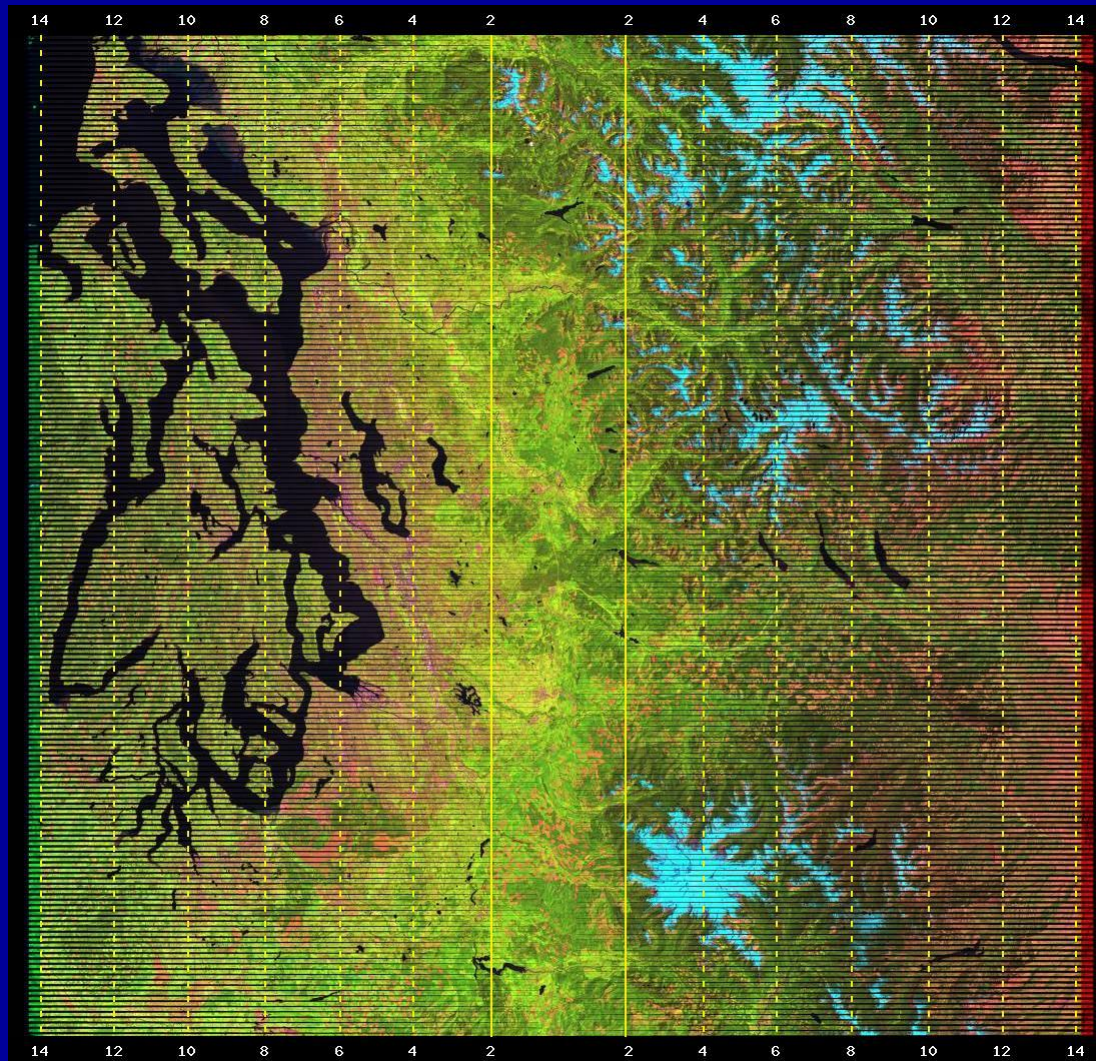


# Gap Masks



# Position of NCCN parks

MORA





# Position of NCCN parks

NOCA

